

## Log: Week 8

This week I have:

- Implemented basic/ placeholder genetic operators
- These allow the algorithm to accurately learn/ approximate straight line routes from  $A \rightarrow B$

Below you can see the population after 1 generation:

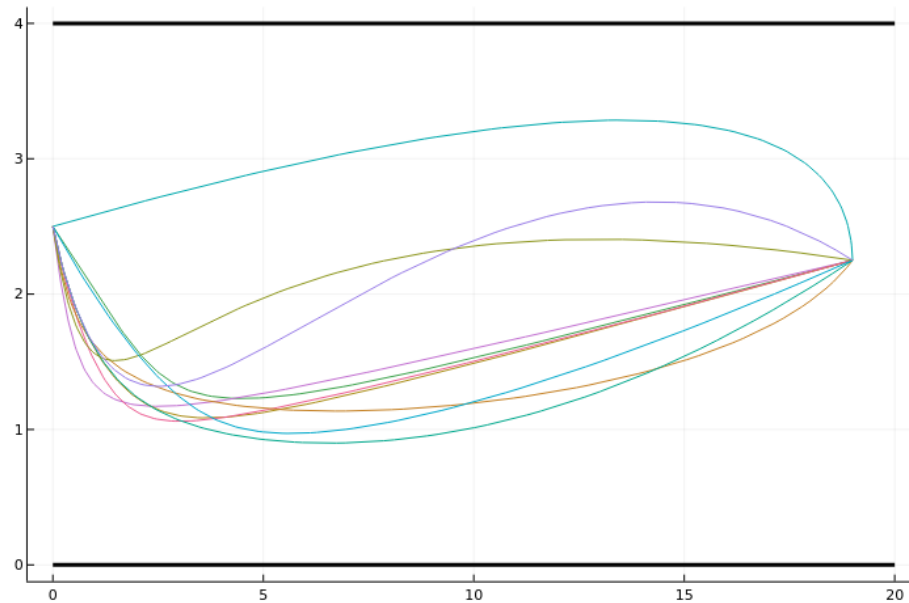


Figure 1: 1 Generation

And the most and least fit solutions after 4 generations

Whilst in isolation this is not particularly impressive, in fact it underperforms a simple Pythagorean distance calculation, it can be generalised and extended to avoid obstacles and to not intersect with other routes.

I went on to implement road space obstacles in the form of circles. I have implemented the abstract type of `Obstacle` of which `Circle` is a member allowing me to extend my program.

By calculating infeasible route sections as the distance between 2 intersects between an individual and an obstacle, I can penalise solutions where such a distance is non-zero.

The results can be seen in Figure 3. These results are seen after just a single

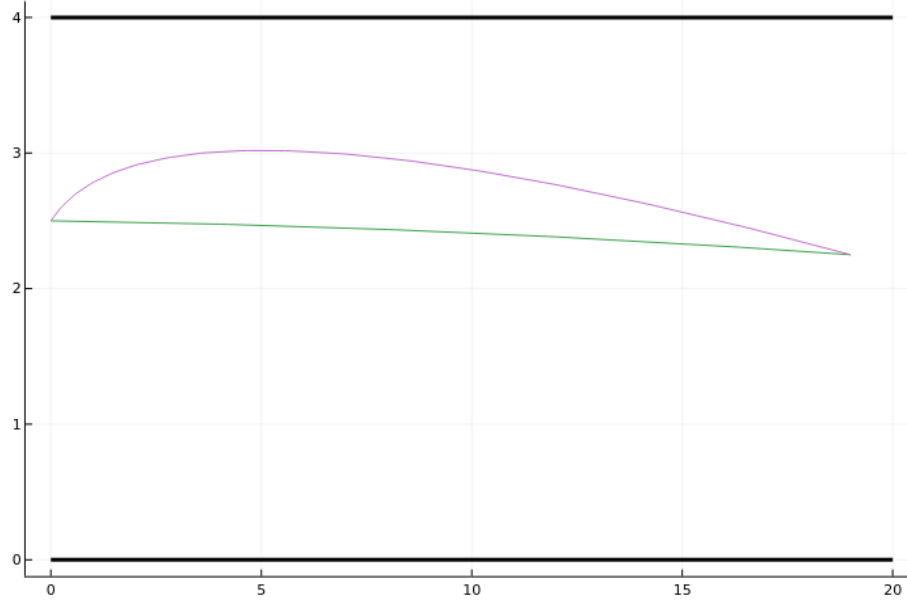


Figure 2: 4 Generations min max

generation with the best solution having approximately 8% lower fitness than the Pythagorean route.

I have now moved on to initial parameter tweaking in order to get the best performance from these basic operators. I have been testing different weightings for infeasible and high proximity routes, I need to make sure they are never the best solution but not penalise them too much so that solutions near them in the search space are never explored; the optimal solution could very well be close to an infeasible one.

You can find a gif of the algorithm over the course of 100 generations [here](#). The bold line is the fittest and the control points for the best curve are shown. Obstacles are shown in red.

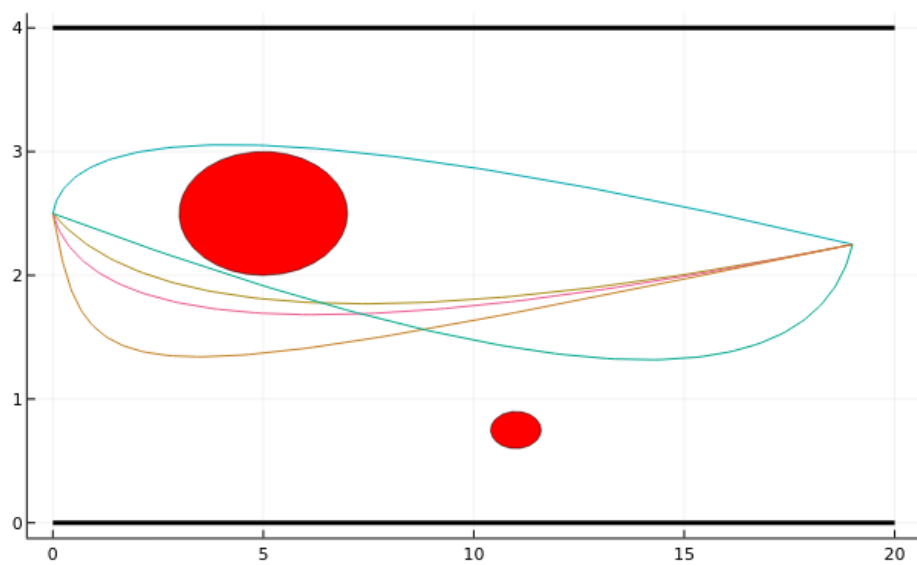


Figure 3: Obstacle avoidance